

Electromagnetic inverse scattering problem

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Green Representation for Electromagnetic field :

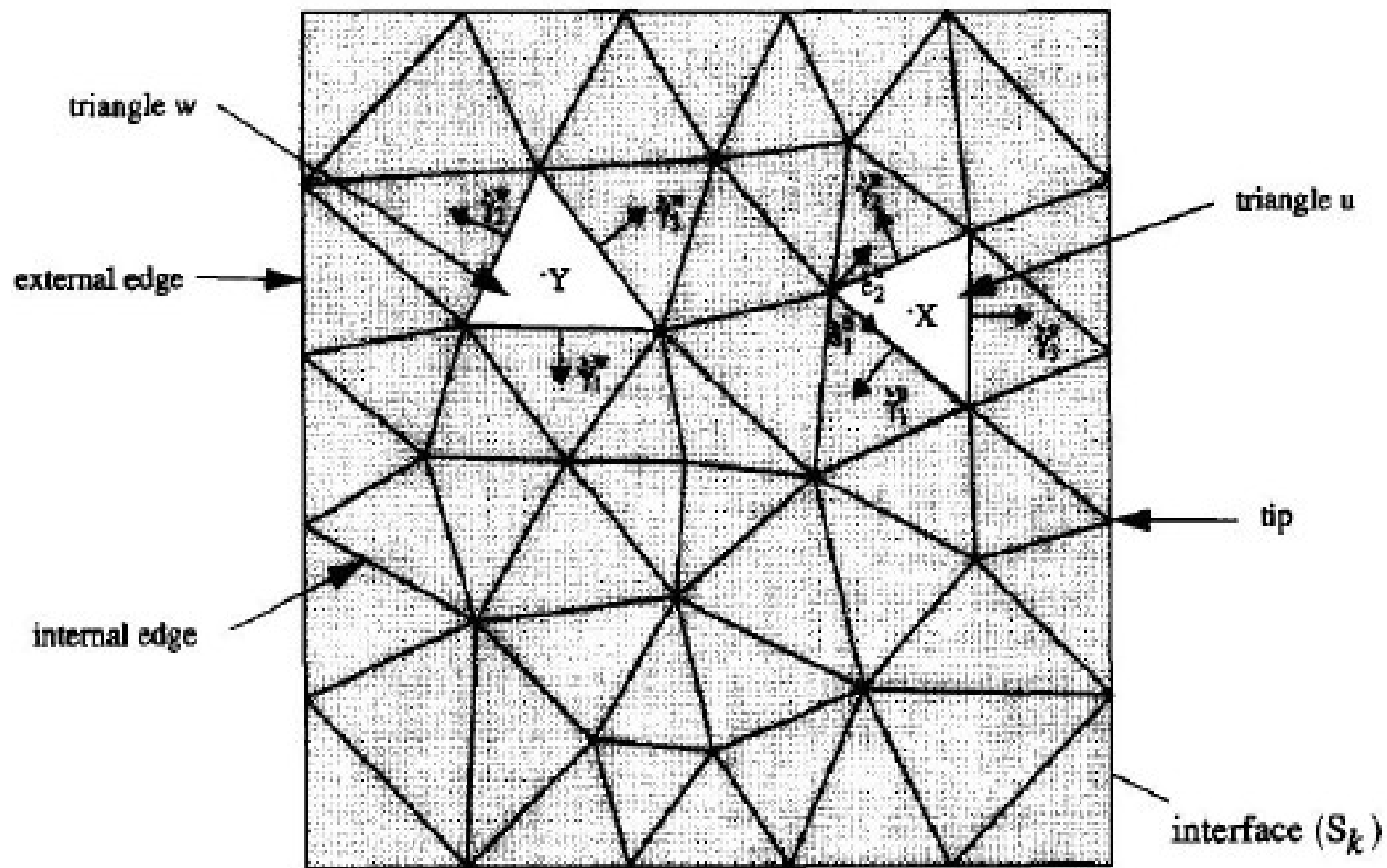
$$\vec{E}_k(x) = \vec{E}_k^i + i\omega\mu \oint_{(S_k)} G(x,y) \vec{j}_k(y) ds(y) - \frac{1}{i\omega\epsilon} \oint_{(S_k)} \text{grad}_x G(x,y) \text{div}_{S_k} \vec{j}_k(y) ds(y) - \text{rot} \oint_{(S_k)} G(x,y) \vec{m}_k(y) ds(y)$$

$$\vec{H}_k(x) = \vec{H}_k^i + i\omega\epsilon \oint_{(S_k)} G(x,y) \vec{m}_k(y) ds(y) - \frac{1}{i\omega\mu} \oint_{(S_k)} \text{grad}_x G(x,y) \text{div}_{S_k} \vec{m}_k(y) ds(y) - \text{rot} \oint_{(S_k)} G(x,y) \vec{j}_k(y) ds(y)$$

Boundary conditions:

$$\vec{n}_k \times \vec{E}_k + \vec{n}_l \times \vec{E}_l = 0 = \vec{m}_k + \vec{m}_l$$

$$\vec{n}_k \times \vec{H}_k + \vec{n}_l \times \vec{H}_l = 0 = \vec{j}_k + \vec{j}_l$$



Rumsey's reaction concept

- The reaction concept between $\vec{J}_k^i \vec{m}_k^i$ and $\vec{J}_k^t \vec{m}_k^t$ is defined by:

$$R_{S_j}(\vec{J}_k^i \vec{m}_k^i, \vec{J}_k^t \vec{m}_k^t) = \oint_{S_j} \vec{E}_j(x) \cdot \vec{J}_k^t(x) - \vec{m}_k^i(x) \cdot \vec{H}_j(x) ds^t(x)$$

- This concept can be applied to $\{\vec{E}_j^i, \vec{H}_j^i\}$ if sources are present in Ω_j :

$$R_{S_j}(\vec{J}_l^a \vec{m}_l^a, \vec{J}_k^t \vec{m}_k^t) = \oint_{S_j} \vec{E}_j^i(x) \cdot \vec{J}_k^t(x) - \vec{m}_k^i(x) \cdot \vec{H}_j^i(x) ds^t(x)$$

Variational formulation of integral equations

- Using the boundary conditions, the total reaction on all domains is expressed as:

$$\sum_{l=1}^N R_{S_l}(\vec{J}_l^a \vec{m}_l^a, \vec{J}_k^t \vec{m}_k^t) + R_{S_j}(\vec{J}_l^a \vec{m}_l^a, \vec{J}_k^t \vec{m}_k^t) = 0$$

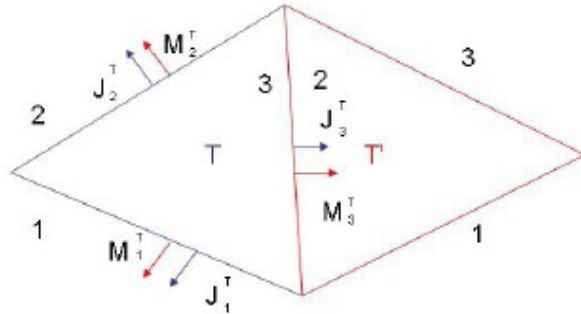
Variational formulation of our problem

$$\sum_{l=1}^N \mu_{rl} Q_{Sl}(S_t, j_l, j_l^t) + \frac{k_l^2}{\mu_{rl}} Q_{sl}(S_t, m_l, m_l^t) - P_{sl}(S_t, j_l, m_l^t) - P_{sl}(S_t, m_l, j_l^t) = - \sum_{l=1}^N \oint_{S^l} (E_l^i(x) j_l^t - H_l^i(x) m_l^t) ds(x)$$

where P_{sl} and Q_{sl} are defined as :

$$Q_{Sl}(S_t, j_l, j_l^t) = \oint_{S^t} \oint_{S^l} G(k, x, y) (j(y) j^t(x) - \frac{1}{k_l^2} \text{div}_S j(y) \text{div}_{S^t} j^t(x)) ds(x) ds^t(y)$$

$$P_{Sl}(S_t, j_t, m_l^t) = \oint_{S^t} \oint_{S^l} [\text{grad}(G(k, x, y) \times j(y))] p^t(x) ds(y) ds^t(x)$$



$$-J_2^{T'} = J_3^T = J_k$$

Currents of the triangle T can be expressed as it is written:

$$\vec{j}^T = \vec{j}_1^T \cdot J_1^T + \vec{j}_2^T \cdot J_2^T + \vec{j}_3^T \cdot J_3^T$$

$$\vec{m}^T = M_1^T \cdot \vec{m}_1^T + M_2^T \cdot \vec{m}_2^T + M_3^T \cdot \vec{m}_3^T$$

Bilinear form is transformed into discrete one :

$$\sum_{l=1}^N \sum_{u=1}^{NE_1} \sum_{w=1}^{NE_1} \mu_{rl} A_l^{uw}(\vec{j}_l, \vec{j}_l) + \frac{k_l^2}{\mu_{rl}} A_l^{uw}(\vec{p}_l, \vec{p}_l) - B_l^{uw}(\vec{j}_l, \vec{j}_l) - B_l^{uw}(\vec{p}_l, \vec{p}_l) = \sum_{l=1}^N \sum_{w=1}^{NE_1} C_l^{rw}(\vec{E}_l, \vec{H}_l, \vec{j}_l, \vec{p}_l)$$

Matrix Formulation of problem : $Ax=C$

Integral solver in comparison with Differential solvers :

Matrix A is dense , contains $N*N$ block

Much more storage space

More CPU time

Hermite interpolation

Hermite (Osculating) polynomials are a generalization of both the Taylor and Lagrangian polynomials. If we have given $n+1$ points x_0, x_1, \dots, x_n and nonnegative integers m_0, \dots, m_n , a Hermite polynomial approximating a function f is a polynomial of at least degree m_i at point x_i where f belongs to $C^m(a, b)$ and $m = \max\{m_0, \dots, m_n\}$ and x_i belongs to $[a, b]$ for every $i = 0, \dots, n$.

The degree of this osculating polynomial will be at most :

$$M = \sum_i m_i + n.$$

The number of conditions to be satisfied is $\sum_i m_i + (n + 1)$ and a polynomial of degree M has $M+1$ coefficients that has to satisfy these conditions.

By definition, if we have $n+1$ distinct numbers in range $[a, b]$ and m_i are nonnegative integers associated with x_i , so for each $i = 0, \dots, n$.

If $m = \max_{0 \leq i \leq n} m_i$ and $f \in C^m[a, b]$

so Hermite polynomial approximating f is the polynomial P of least degree m such that

$$\frac{d^k P(x_i)}{dx^k} = \frac{d^k f(x_i)}{dx^k} \text{ for each } i = 0, 1, \dots, n \text{ and } k = 0, 1, \dots, m_i.$$

If the function f belongs to $C^1[a, b]$ and x_0, x_1, \dots, x_n that are in range $[a, b]$ are distinct, the unique polynomial of least degree matching with f and f' at x_0, x_1, \dots, x_n is the polynomial of degree at most $M=2n+1$ given as :

$$H_{2n+1}(x) = \sum_j f(x_j) H_{n,j}(x) + \sum_j f'(x_j) \hat{H}_{n,j}(x)$$

where :

$$H_{n,j}(x) = [1 - 2(x - x_j)L'_{n,j}(x_j)]L_{n,j}^2(x)$$

and

$$\hat{H}_{n,j}(x) = (x - x_j)L_{n,j}^2(x)$$

here $L_{n,j}(x)$ denotes Lagrange coefficient polynomial of degree n that is defined by formula :

$$L_{n,k}(x) = \frac{(x - x_0) \dots (x - x_{k-1})(x - x_{k+1}) \dots (x - x_n)}{(x_k - x_0) \dots (x_k - x_{k-1})(x_k - x_{k+1}) \dots (x_k - x_n)}$$

k-th divided difference with respect to $x_i, x_{i+1}, \dots, x_{i+k}$ will be :

$$f[x_i, x_{i+1}, \dots, x_{i+k}] = \frac{f[x_{i+1}, \dots, x_{i+k}] - f[x_i, x_{i+1}, \dots, x_{i+k-1}]}{x_{i+k} - x_i}$$

Using Newton divided differences we can simplify programming of interpolation

$$E(\varepsilon) = c_5\varepsilon^5 + c_4\varepsilon^4 + c_3\varepsilon^3 + c_2\varepsilon^2 + c_1\varepsilon + c_0$$

Newton divided differences

x	f(x)	First divided differences	Second divided differences	Third divided differences
x_0	$f[x_0]$			
		$f[x_0, x_1] = \frac{f[x_1] - f[x_0]}{x_1 - x_0}$		
x_1	$f[x_1]$		$f[x_0, x_1, x_2] = \frac{f[x_1, x_2] - f[x_0, x_1]}{x_2 - x_0}$	
		$f[x_1, x_2] = \frac{f[x_2] - f[x_1]}{x_2 - x_1}$		$f[x_0, x_1, x_2, x_3] = \frac{f[x_1, x_2, x_3] - f[x_0, x_1, x_2]}{x_3 - x_0}$
x_2	$f[x_2]$		$f[x_1, x_2, x_3] = \frac{f[x_2, x_3] - f[x_1, x_2]}{x_3 - x_1}$	
		$f[x_2, x_3] = \frac{f[x_3] - f[x_2]}{x_3 - x_2}$		$f[x_1, x_2, x_3, x_4] = \frac{f[x_2, x_3, x_4] - f[x_1, x_2, x_3]}{x_4 - x_1}$
x_3	$f[x_3]$		$f[x_2, x_3, x_4] = \frac{f[x_3, x_4] - f[x_2, x_3]}{x_4 - x_2}$	
		$f[x_3, x_4] = \frac{f[x_4] - f[x_3]}{x_4 - x_3}$		$f[x_0, x_1, x_2, x_3] = \frac{f[x_3, x_4, x_5] - f[x_2, x_3, x_4]}{x_5 - x_2}$
x_4	$f[x_4]$		$f[x_3, x_4, x_5] = \frac{f[x_4, x_5] - f[x_3, x_4]}{x_5 - x_3}$	
		$f[x_4, x_5] = \frac{f[x_5] - f[x_4]}{x_5 - x_4}$		
x_5	$f[x_5]$			

z	$f(z)$	First divided differences	Second divided differences
$z_0 = x_0$	$f[z_0] = f[x_0]$		
		$f[z_0, z_1] = f'(x_0)$	
$z_1 = x_0$	$f[z_1] = f[x_0]$		$f[z_0, z_1, z_2] = \frac{f[z_1, z_2] - f[z_0, z_1]}{z_2 - z_0}$
		$f[z_1, z_2] = \frac{f[z_2] - f[z_1]}{z_2 - z_1}$	
$z_2 = x_1$	$f[z_2] = f[x_1]$		$f[z_1, z_2, z_3] = \frac{f[z_2, z_3] - f[z_1, z_2]}{z_3 - z_1}$
		$f[z_2, z_3] = f'(x_1)$	
$z_3 = x_1$	$f[z_3] = f[x_1]$		$f[z_2, z_3, z_4] = \frac{f[z_3, z_4] - f[z_2, z_3]}{z_4 - z_2}$
		$f[z_3, z_4] = \frac{f[z_4] - f[z_3]}{z_4 - z_3}$	
$z_4 = x_2$	$f[z_4] = f[x_2]$		$f[z_3, z_4, z_5] = \frac{f[z_4, z_5] - f[z_3, z_4]}{z_5 - z_3}$
		$f[z_4, z_5] = f'(x_2)$	
$z_5 = x_2$	$f[z_5] = f[x_2]$		

$$H_5 = f[z_0] + \sum_{k=1}^{2n+1} f[z_0, \dots, z_k](x-z_0)(x-z_1)\dots(x-z_{k-1})$$

$$H(x) = Q_{0,0} + (x - x_0)Q_{1,1} + Q_{2,2}(x - x_0)^2 + Q_{3,3}(x - x_0)(x - x_1) + Q_{4,4}(x - x_0)^2(x - x_1)^2 + Q_{5,5}(x - x_2)(x - x_0)^2(x - x_1)^2$$

This polynomial is transformed in following form:

$$E(\varepsilon) = c_5\varepsilon^5 + c_4\varepsilon^4 + c_3\varepsilon^3 + c_2\varepsilon^2 + c_1\varepsilon + c_0$$

Searching for all possible values for which $P(\epsilon)=0$

Searching for zeros is equivalent to problem where we want to calculate eigenvalues of companion matrix :

$$A = \begin{pmatrix} -\frac{c_4}{c_5} & -\frac{c_3}{c_5} & -\frac{c_2}{c_5} & -\frac{c_1}{c_5} & -\frac{c'}{c_5} \\ 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \end{pmatrix}$$

QR method for computing eigenvalues

- Gram-Schmidt
- Given rotation
- Householder transformations
- ...

Classical Gram -Shmidt

$$u_1 = a_1 \text{ and } q_1 = \frac{u_1}{\|u_1\|}$$

$$u_k = a_k - \sum_{j=1}^{k-1} \text{proj}_{u_j} a_k \qquad q_k = \frac{u_k}{\|u_k\|}$$

In k-th step we first decompose A and then we calculate A in next iteration

$$A^k = Q * R$$

$$A^{k+1} = R * Q$$

After certain number of iteration we will have convergence

Modified Gram-Schmidt

```
do k=1,5
```

```
  r(k,k)=sqrt(dot_product(A(1:5,k),A(1:5,k)))
```

```
  q(1:5, k) = A(1:5, k) / r(k,k);
```

```
    do j = k+1,5
```

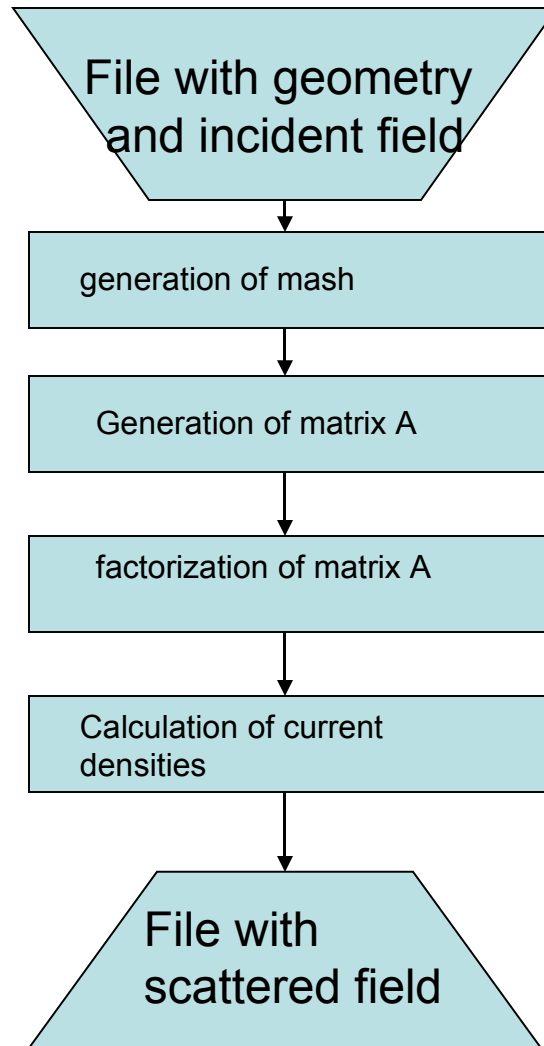
```
      r(k, j) = dot_product(q(1:5, k), A(1:5, j));
```

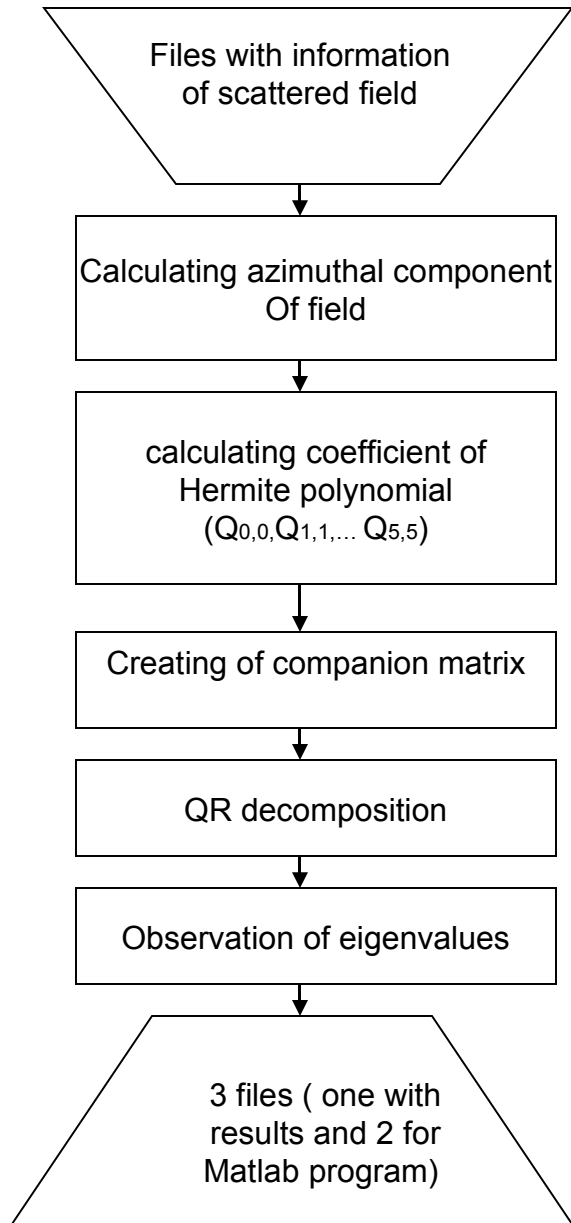
```
      A(1:5, j) = A(1:5, j) - r(k, j) * q(1:5, k);
```

```
    end do
```

```
end do
```

Structure of SR3D





Observation of eigenvalues

We have to check if we have 2 X 2 blocks and to flag corresponding elements

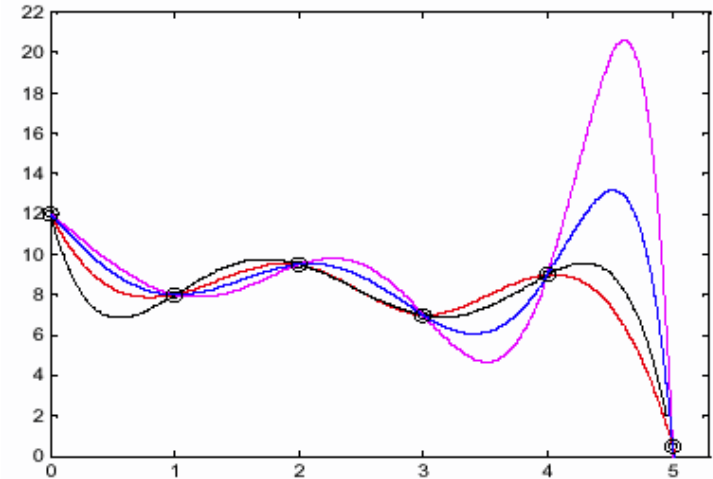
Checking if real eigenvalues are in corresponding range

If its not flaged and its in range (min ϵ , max ϵ) it is written in file results.txt

Various ways for approximation of function

If only values of function are known:

Least-square method,
Polynomial and trigonometric functions,
Taylor series and Chebyshev approximations,
Piecewise polynomial functions,
Splines, cubic splines, B-splines functions,
Rational functions, Padé approximations,
Thiele interpolations,
FFT, Neural networks, Genetic algorithms
and so on.



If we have data about values
and derivatives of function:

Splines, cubic splines, B-splines ...
Piecewise polynomial functions
Neural networks, Genetic
algorithms